

# High-resolution ab-initio three-dimensional coherence X-ray diffraction microscopy

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June 20, 2005

Coherence 2005 Porquerolles, France June 14, 2005 through June 18, 2005

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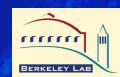
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## High-resolution ab initio three-dimensional coherent X-ray diffraction imaging

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#### Overview

- Critical issues in obtaining a high-quality image include:

  Data collection: signal to noise, system stability, dynamic range, automation
  Alignment of diffraction patterns with respect to one another
  Assembly of the diffraction data into a diffraction volume

  - Efficient algorithms for applying phase retrieval techniques to the diffraction volume
     Stability of the three-dimensional phase retrieval process

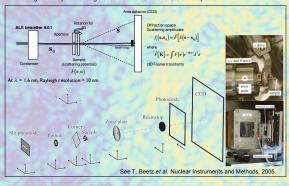
  - Techniques for determining the object support; and
     Treatment of missing data, both within the beamstop region and elsewhere.

We have obtained high-quality 3D reconstructions from X-ray diffraction data alone. This is an important step, as it does not require a low-resolution image to fill in the beamstop

#### Data collection

- Data collected at beamline 9.0.1 at the Advanced Light Source (ALS), Berkeley using an end-station built by Chris Jacobsen's team at State University of New York at Stony
- Brook. The beamline delivers a coherent flux of 8x10<sup>9</sup> photons/<mark>sec</mark> into <mark>a 4μm pinhole with Δλ/</mark>
- λ of several thousand using an off-axis zone plate / pinhole pair monochromator.

  Sample positioning uses a motorised four-axis goniometer system provided by JEOL USA, Inc., which is normally a component of their JEM-2010 FasTEM transmission electron princescope.
- In-vacuum CCD camera from Roper Scientific has a backside-illuminated 1340x1300
- In-vacuum CCD camera from Noper Scientific rises to usus-assuminations to Technologies (CCD with 20µm pixel SCD with 20µm pixels and a claimed readout noise of 4 electrons RMS per pixel. Dynamic range of the CCD is limited to 200000 electrons full well capacity, so several exposures ranging from 0.5sec to 100sec are merged in software to produce a single image with dynamic range several times that of the CCD chip itself.

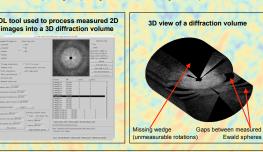


#### Data assembly and preprocessing

The measured 2D diffraction patterns must first be converted into a 3D data cube:

- Assembly of the 3D diffraction volume is performed by mapping the 2D (x,y) co-ordinates of each pixel in the measured diffraction pattern onto the Ewald sphere using the known distance from sample to detector, beam energy and CCD pixel size and to determine the Ewald sphere curvature.
- A 3D rotation matrix from the known sample orientation is then applied to these coordinates to determine the 3D coordinates of each measured pixel in the
- The rotated Ewald sphere diffraction patterns are mapped onto a regular 3D cartesian voxel grid using nearest-neighbour sampling.

  Where more than one pixel from the set of intensity measurements contributed to a given voxel, the pixel values are averaged to determine the appropriate intensity value at that point.
- Only the measured diffraction data are used no other sample information or measurements were used in reconstruction process.
- No low-resolution measurements are used to fill in the beamstop region and no numerical interpolation is performed between the measured Ewald spheres.
- For the case of the pyramid data set (shown below) 149 diffraction patterns were measured spanning sample angles from -45 to +78 degrees sample rotation.





#### **Computational reconstruction**

Each reconstruction requires at least

- Two complex arrays (reconstructions)
  One floating point array (input data)
  One byte mask (support)
  1000 or more 3D FFTs

significant and suggest a cluster-based solution

256<sup>3</sup> 336MB 592MB 512<sup>3</sup> 2.6GB 4.7Gb 1024<sup>3</sup> 22GB 38GB 176GB

We use the dist\_fff distributed fast Fourier transform library from Apple Computer for optimum Fourier transformn speed. dist\_fft has been hand-optimised for the G5 vector processor architecture by the Apple Advanced Computation Group and uses standard MPI interfaces to perform distributed giga-element or larger FFTs.

Reconstruction code is written in C, is fully parallelised, and uses distributed memory and MPI interfaces to share the workload across all CPUs in the system. This includes application of real and Fourier space constraints and dynamic support refinement using the Shrinkwap algorithm.

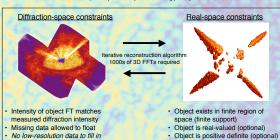
We use a 16-node 2 0GHz dual processor (32 processors total)
Xserve G5 cluster with low latency
Infiniband interconnects and 4GB

RAM per node.

Reconstructions presented here were performed on a 512<sup>3</sup> grid enabling us to produce a high-quality reconstruction in 1.5 to 2 hours.

	Reconstruction speed on 16-node, 32-CPU 2.0GHz Xserve G5 cluster		
	Size	Per 3D FFT	Full reconstruction <sup>2</sup>
	256³	73msec	10 mins
	512 <sup>3</sup>	850msec	1.5 hrs
	1024³	7.9sec	14 hrs

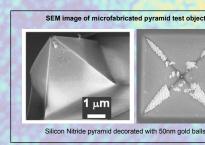
<sup>1</sup> Assumes minimum of 2x complex data cub es for FFTs, 1x diffraction data cube (real), 1x support array (byte)
<sup>2</sup> Based on 2000 iterations, 2FFTs per iteration plus other floating point operations needed for the reconstruction

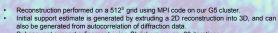


### 3D images from experimental data

- Test object consists of a microfabricated silicon nitride pyramid decorated with 50nm gold balls, as shown in the SEM images below.

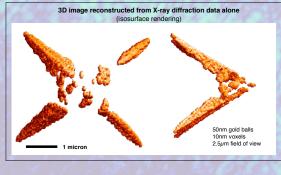
  149 diffraction patterns were measured spanning sample angles from -45 to +78 degrees in steps of 0.5 to 1.0 degrees at an X-ray energy of 750eV. Individual diffraction images are arranged into a 3D diffraction cube using known sample rotation angles. The assembled 3D diffraction volume has been shown above.

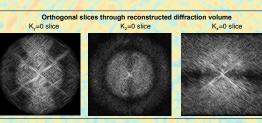




Subsequent support refinement using *Shrinkwrap* every 30 iterations. Support threshold of 15%, initial convolution radius of 1.2 pixels Gaussian FWHM, reducing to 0.75 pixels in steps of 2% per application.

HIO algorithm used for first 600 iterations, then RASR algorithm to iteration 2500.





Missing data (including the beamstop region) are allowed to float. Regions of missing data are filled in by the reconstruction code, producing a continuous diffraction pattern representing the object.

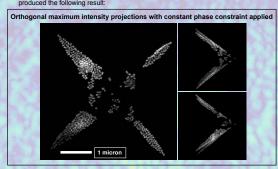
#### Application of material constraints

s, the reconstructed scattering potential takes the form: 
$$F(\mathbf{r},\omega) = \frac{1}{4\pi} k^2 \left[ n^2 (\mathbf{r},\omega) - 1 \right]$$

$$n = 1 - \delta - i\beta$$

$$F(\mathbf{r},\omega) = \frac{1}{2\pi} k^2 \left[ -\delta - i\beta + O(\delta,\beta)^2 \right]$$

- For a single material the ratio of  $\delta$  and  $\beta$  is constant, thus all voxels should have the
- composed of the linear sum of all  $\delta,\beta$  values for all materials in the object.
- Application of the single material phase constraint to our gold ball pyramid data produced the following result:



#### 3D imaging of low density foams

We have applied 3D diffraction imaging to determining the 3D structure of low density aerogel foam samples. These foams are low density (10mg/cc) and have an internal skeleton structure composed of Ta<sub>2</sub>O<sub>5</sub>.

Data collection and processing is similar to that used for the pyramid object. For ease of handling the sample was placed on a silicon nitride membrane which collected some debris (as shown in reconstruction to the left).

Scientific question is: How isotropic and uniform are these foams, and how does processing alter their structure?

3D structure measured using diffraction imaging is being used to answer this question.

